HiHAT: A Way Forward to Perf Portability with Retargetable Infrastructure

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OUTLINE

- Perspective: performance portability
- Challenges: More heterogeneity in HW platforms, SW interfaces
- Solutions: Common retargetable infrastructure hierarchical hetero async tasking

HETEROGENEITY AND RETARGETABILITY

- Heterogeneity within a platform
 - Increasing specialization
 - Host, accelerators; kinds, layers and locations of memory; interconnect
- Retargetability across platforms
 - One software architecture, many targets
 - And of course we want...

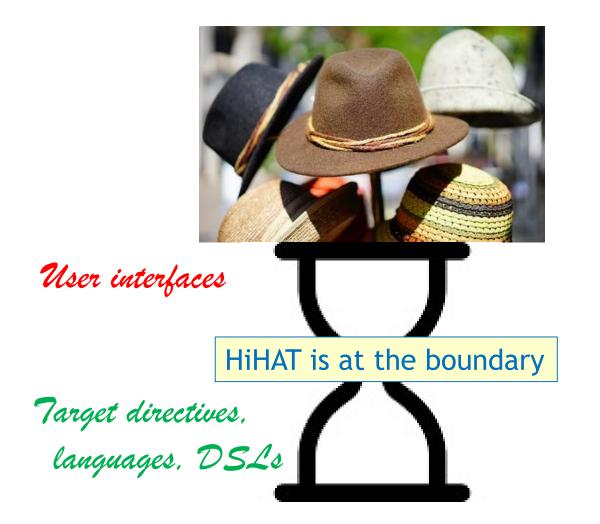
PERFORMANCE PORTABILITY DEFINITION

"Same code" + different architectures → efficient performance

PERFORMANCE PORTABILITY CONTRADICTIONS

• "Same code" + different architectures → efficient performance

- Contradictions first set
 - But I like my language! The other guy's language gives horrible performance!
 - But I need a special data layout for each target!
 - But I have a favorite user-level interface. Don't take that away from me!



target agnostic

target specific

PERFORMANCE PORTABILITY PARTIAL SOLUTIONS

• "Same code" + different architectures → efficient performance

- Potential solutions first set
 - Language: Target-specific task implementations where needed
 - Data layout: Task implementations tailored for data layout, scheduler can choose to relayout data off of the critical path
 - User-level interface: Layer client user-facing runtimes on top of retargetable interface

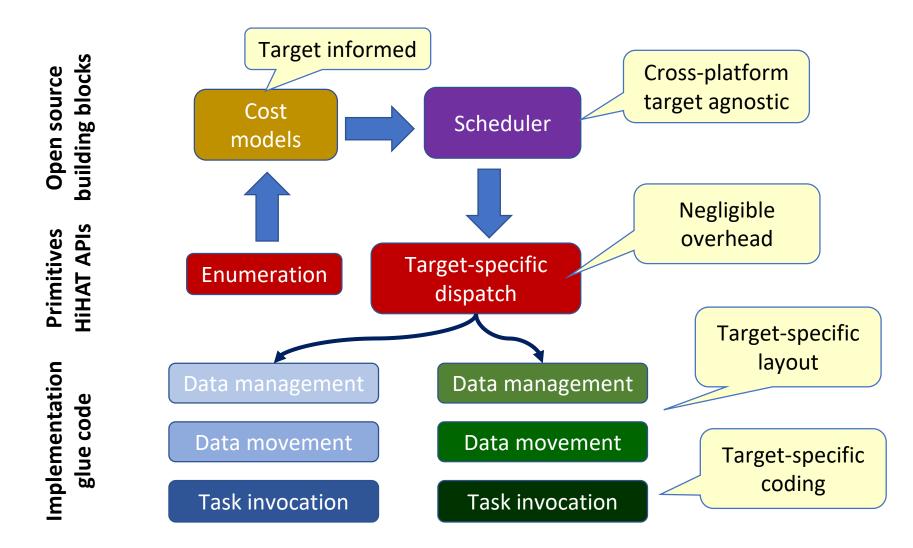
Task: High-level language, with directives or DSL or even assembly instructions

- Pluggable implementations
 - Task: High-level language, with directives or DSL or even assembly instructions
 - Best way for a given platform: target-specific APIs and implementations

- Sequence of target-agnostic primitives
 - Invoke, manage data, move data, coordinate, enumerate
- Pluggable implementations
 - Task: High-level language, with directives or DSL or even assembly instructions
 - Best way for a given platform: target-specific APIs and implementations

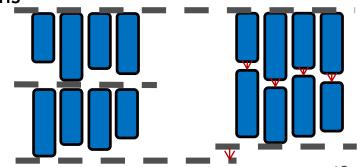
- Scheduler binding and ordering, based on cost model
 - Select target, implementation, layout, add actions as needed
 - Invoke primitives where and when most appropriate
- Sequence of target-agnostic primitives
 - Invoke, manage data, move data, coordinate, enumerate
- Pluggable implementations
 - Task: High-level language, with directives or DSL or even assembly instructions
 - Best way for a given platform: target-specific APIs and implementations

COMMON RETARGETABLE SW ARCHITECTURE



MOTIVATIONS FOR A SCHEDULER

- Lack of predictability
 - Where data comes from, in memory hierarchy or across network
 - When computation will finish: complex algorithms, load imbalance, DVFS
- Growing complexity
 - Too many factors at play to settle on a single portable static scheduler
 - Too much diversity in increasingly-heterogeneous platforms
- Going asynchronous
 - Break out of bulk synchronous, move to point-point
 - Dynamic management of resources

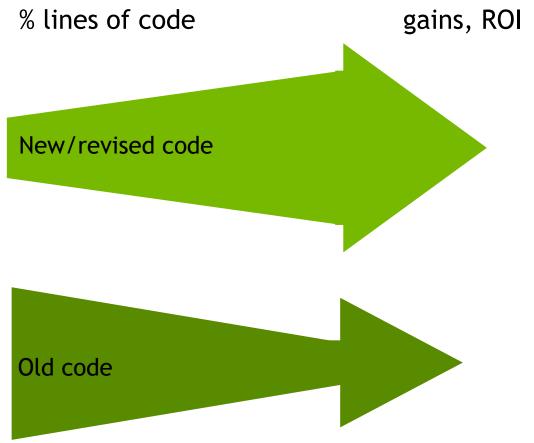


PROVIDING ACCESS TO PERFORMANCE

Meeting our customers where they are, offering a path forward

- Exposing maximal parallelism
- Extreme scaling
- Tuning for the target platform
- Tailored abstractions

- Limited effort
- Traditional language interfaces

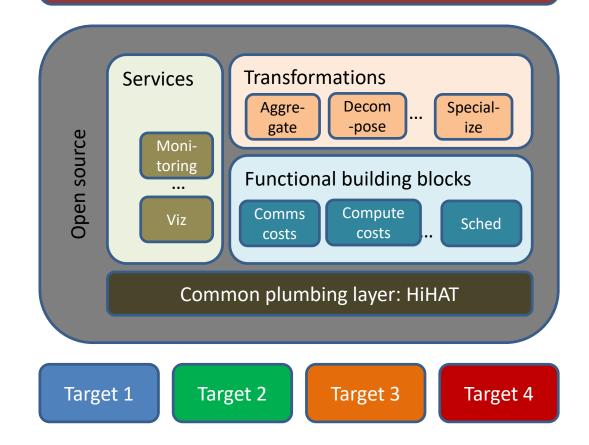


App developers code

Applications and frameworks: compilers, runtime libraries, ...

Tuners configure

Experts implement



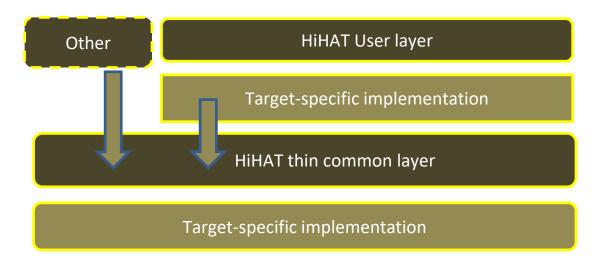
https://wiki.modelado.org/Heterogeneous_Hierarchical_Asynchronous_Tasking

HIHAT: APIS FOR RETARGETABILITY

- Plug in target-specific implementations from below
- Implement data management, data movement, invocation, coordination, querying

User: ease of use via abstraction

Common: minimal overhead



LANGUAGE OR TASKING FRAMEWORKS

Some part of each institution has expressed technical interest, not necessarily business commitment.

- C++ (CodePlay, IBM) Michael Wong
- Chapel (Cray), Brad Chamerlain
- Charm++ (UIUC) Ronak Buch, (Charmworks)
 Phil Miller
- Darma (**Sandia**) Janine Bennett
- Exa-Tensor (ORNL) Wayne Joubert
- Gridtools (CSCS, Titech) Mauro Bianco
- HAGGLE (PNNL/HIVE) Antonino Tomeo
- Kokkos, Task-DAG (SNL) Carter Edwards
- Legion (Stanford/NV) Mike Bauer
- OmpSs (BSC) Jesus Labarta

- Realm (Stanford/NV) Sean Treichler
- OCR (Intel, Rice, GA Tech) Vincent Cave
- PaRSEC (UTK) George Bosilca
- Raja (LLNL) Rich Hornung
- Rambutan, UPC++ (LBL) Cy Chan
- R-Stream (Reservoir Labs) Rich Lethin
- StarPU (INRIA) Samuel Thibault
- SyCL (CodePlay) Michael Wong
- SWIFT (Durham) Matthieu Schaller
- TensorRT (NVIDIA) Dilip Sequeira
- VMD (UIUC) John Stone

TABULATED RESULTS

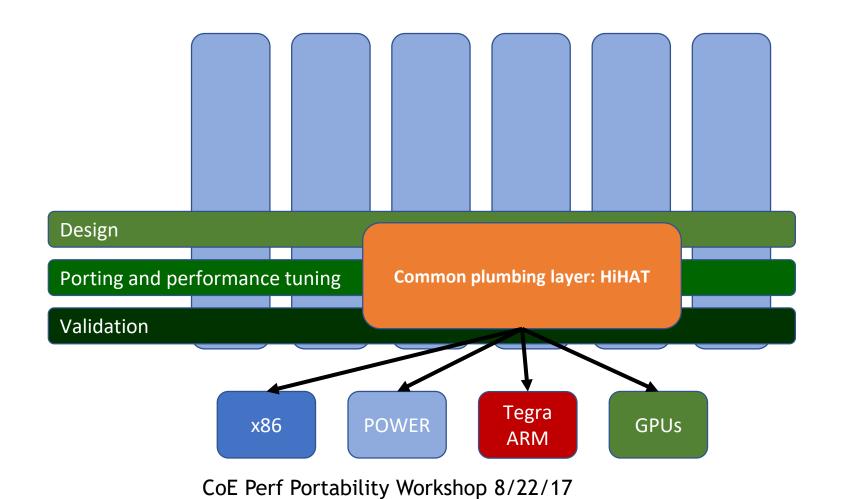
Strong interest, modestly amenable; progress; next

Type of functionality	Level of interest		Amenability to refactoring			
	Н	M	L	Н	M	L
Data movement - target-optimized copies, DMA, networking	15	1	1	7	5	1
Data management - kinds and layers of memory, specialized pools	11	4	2	7	4	2
Coordination - completion events, locks, queues, collectives, iteration	9	8	0	6	5	1
Compute - local or remote invocation	7	3	4	4	5	4
Enumeration - kinds/# of resources, topologies	11	5	1	4	4	3
Feedback - profiling, utilization	6	7	2	4	7	1
Tools - tracing, callbacks, pausing, debugging	3	12	2	2	7	2

ADOPTION

- Meet requirements
 - Provisioning: C ABI, library, interoperable, profiling
 - Performance: enables access to perf features, low overhead \rightarrow supports fine granularity
 - Incremental, easy on ramp
- Open architecture
 - Be a provider for tasking and language runtimes and frameworks
 - Plug in implementations from below, from vendors or third parties
 - Share building blocks, e.g. cost models, schedulers
- Easiest and best solution

SO MANY FRAMEWORKS, SO LITTLE TIME



PROTOTYPE INFRASTRUCTURE CAPABILITIES

The basics are already working

- Current test platform: 2 CPU sockets + 2 GPUs in one node
- Data movement
 - User Layer: <dst, src, size> using logical handles for addressing
 - Common Layer: use specialized flavors
 - Set up comms, establish visibility as needed
- Data management
 - User Layer: Allocate or register, and create address-memory resource association
 - Also support tagging to clean up a set of allocations/wraps at once
 - Common Layer: No tagging
- Invocation
 - Register target-specific implementations, invocation with closure
- Microbenchmarks show overheads are within measurement noise
 CoE Perf Portability Workshop 8/22/17

MOLECULAR ORBITALS (MO) APPLICATION

- Compute wavefunction amplitudes on a grid for visualization
 - Evaluate linear combination of Gaussian contractions (polynomials) at each grid point, function of distance from atoms
- Algorithm made arithmetic bound via fast on-chip memory systems
- Three different algorithms for different memory structures:
 - GPU constant memory
 - Shared memory tiling
 - L1 global memory cache
- Representative of a variety of other grid-oriented algorithms, stencils
- Use of special GPU hardware features, APIs helped drive completeness of HiHAT proof-of-concept implementation already at an early stage

MOLECULAR ORBITALS PERFORMANCE

HIHAT API GAINS FOR MOLECULAR ORBITALS APPLICATION

- Performance of MO algorithm on HiHAT User Layer PoC implementation closely tracks CUDA performance.
- Spans x86, POWER and Tegra ARM CPUs

Molec	ular Orbital Algorithm, Mem Kind	Speedup	HiHAT
		vs. ShMem	API gain
x86	SharedMem HiHAT	1.000x	1.028x
+	L1CachedGlblMem HiHAT	1.088x	1.025x
GPU	ConstMem HiHAT	1.472x	1.031x
PWR	SharedMem HiHAT	1.000x	0.999x
+	L1CachedGlblMem HiHAT	1.116x	1.001x
GPU	ConstMem HiHAT	1.534x	0.983x
ARM	SharedMem HiHAT	1.000x	-
+	L1CachedGlblMem HiHAT	1.094x	-
GPU	ConstMem HiHAT	1.059x	-
	NoPin-SharedMem HiHAT	2.349x	0.995x
	NoPin-L1CachedGlblMem HiHAT	2.561x	0.984x
	NoPin-ConstMem HiHAT	2.562x	0.998x

PORTABILITY ON MO

Mapping between CUDA and HiHAT

- Time to port MO: 90 minutes
- HiHAT has fewer unique APIS (6 vs. 10)
- HiHAT has fewer static API calls (30 vs. 38)
- Accelerate optimization space exploration
- Also enhance coding productivity

TARGET-SPECIFIC API USAGE IN MOLECULAR ORBITALS APPLICATION

Category	Original CUDA		Ported to HiHAT		
Invoke	<<<>>>	3	hhuInvoke()	3	
Data mvt	cudaMemcpy()	7	hhuCopy()	7	
	cudaMemcpyToSymbol()	7	hhuCopy()	2	
Configuration	cudaSetDeviceFlags()	1	(config)	0	
	cudaFuncSetCacheConfig()	2	(config)	0	
Data mgt,	cudaMalloc()	7	hhuAlloc()	7	
minimal	cudaMallocHost()	1	hhuAlloc()	1	
	cudaHostAlloc()	1	hhuAlloc()	1	
	[free]		hhuClean()	[1]	
	[symbols]	-	hhuRegMem()	7	
Data mgt,	cudaFree()	7	hhuFree()	(7)	
eliminatable	cudaFreeHost()	2	hhuFree()	(2)	
	[symbols]	-	hhuDeregMem()	(7)	
Coordination	-	0	hhuSyncAll()	1	
Totals					
static	14+3+3+9+9+0	38	9+3+0+16+16+1	43	
static min'l	14+3+3+9+9+0	38	9+3+0+17+0 +1	30	
unique	2+1+2+5+0+0	10	1+1+0+2+2+1	7	
unique min'l	2+1+2+5+0+0	10	1+1+0+3 +0 +1	6	

TAKE-AWAYS

- Portability comes at the scheduling layer, on top of target-agnostic primitives
- Dynamic scheduling may have the most promising path to portability and scaling
- Necessary conditions: meet requirements; be pluggable; open source approach;
 be the easiest path to performance, generality and robustness
- HiHAT prototype looks promising as a retargetable infrastructure